# Non-Life Insurance: Mathematics and Statistics

## Exercise sheet 6

### Exercise 6.1 Log-Normal Distribution and Deductible

Assume that the total claim amount

$$S = \sum_{i=1}^{N} Y_i$$

in a given line of business has a compound distribution with  $\mathbb{E}[N] = \lambda v$ , where  $\lambda > 0$  denotes the claim frequency and v > 0 the volume, and with a log-normal distribution with mean parameter  $\mu \in \mathbb{R}$  and variance parameter  $\sigma^2 > 0$  as claim size distribution.

(a) Show that

$$\mathbb{E}[Y_1] = \exp\left\{\mu + \frac{\sigma^2}{2}\right\},$$
  

$$\operatorname{Var}(Y_1) = \exp\left\{2\mu + \sigma^2\right\} \left(\exp\left\{\sigma^2\right\} - 1\right) \quad \text{and}$$
  

$$\operatorname{Vco}(Y_1) = \sqrt{\exp\left\{\sigma^2\right\} - 1}.$$

- (b) Suppose that  $\mathbb{E}[Y_1] = 3'000$  and  $\operatorname{Vco}(Y_1) = 4$ . Up to now, the insurance company was not offering contracts with deductibles. Now it wants to offer a deductible of d = 500. Answer the following questions:
  - (i) How does the claim frequency  $\lambda$  change by the introduction of the deductible?
  - (ii) How does the expected claim size  $\mathbb{E}[Y_1]$  change by the introduction of the deductible?
  - (iii) How does the expected total claim amount  $\mathbb{E}[S]$  change by the introduction of the deductible?

Exercise 6.2 Akaike Information Criterion and Bayesian Information Criterion Assume that we fit a gamma distribution to a set of n = 1'000 i.i.d. claim sizes and that we obtain the following method of moments (MM) estimates and maximum likelihood estimates (MLE):

$$\hat{\gamma}^{MM} = 0.9794$$
 and  $\hat{c}^{MM} = 9.4249$ ,  
 $\hat{\gamma}^{MLE} = 1.0013$  and  $\hat{c}^{MLE} = 9.6360$ .

The corresponding log-likelihoods are given by

$$\ell_{\mathbf{Y}}\left(\widehat{\gamma}^{\mathrm{MM}}, \widehat{c}^{\mathrm{MM}}\right) = 1'264.013 \text{ and } \ell_{\mathbf{Y}}\left(\widehat{\gamma}^{\mathrm{MLE}}, \widehat{c}^{\mathrm{MLE}}\right) = 1'264.171.$$

- (a) Why is  $\ell_{\mathbf{Y}}(\widehat{\gamma}^{MLE}, \widehat{c}^{MLE}) > \ell_{\mathbf{Y}}(\widehat{\gamma}^{MM}, \widehat{c}^{MM})$ ? Which fit should be preferred according to the Akaike Information Criterion (AIC)?
- (b) The estimates of  $\gamma$  are very close to 1 and, thus, we could also use an exponential distribution as claim size distribution. For the exponential distribution we obtain the MLE  $\hat{c}^{\text{MLE}} = 9.6231$ and the corresponding log-likelihood  $\ell_{\mathbf{Y}}(\hat{c}^{\text{MLE}}) = 1'264.169$ . According to the AIC and the Bayesian Information Criterion (BIC), should we prefer the gamma model or the exponential model?

#### Exercise 6.3 Goodness-of-Fit Test

Suppose we are given the following claim size data (in increasing order) coming from independent realizations of an unknown claim size distribution:

 $210,\,215,\,228,\,232,\,303,\,327,\,344,\,360,\,365,\,379,\,402,\,413,\,437,\,481,\,521,\,593,\,611,\,677,\,910,\,1623.$ 

(a) Use the intervals

 $I_1 = [200, 239), \quad I_2 = [239, 301), \quad I_3 = [301, 416), \quad I_4 = [416, 725), \quad I_5 = [725, +\infty)$ 

to perform a  $\chi^2$ -goodness-of-fit test at significance level of 5% to test the null hypothesis of having a Pareto distribution with threshold  $\theta = 200$  and tail index  $\alpha = 1.25$  as claim size distribution.

(b) In goodness-of-fit tests with K disjoint intervals and a total of n observations we use the test statistic

$$X_{n,K}^2 = \sum_{k=1}^{K} \frac{(O_k - E_k)^2}{E_k}.$$

where  $O_k$  denotes the actual number of observations and  $E_k$  the expected number of observations in the k-th interval. We assume that the parameters of the null hypothesis distribution function are given and that the K disjoint intervals are chosen such that  $E_k > 0$ , for all  $k = 1, \ldots, K$ . Show that in case of K = 2 disjoint intervals, the test statistic  $X_{n,2}^2$  converges to a  $\chi^2$ -distribution with one degree of freedom, as  $n \to \infty$ .

#### Exercise 6.4 Kolmogorov-Smirnov Test

Suppose we are given the following data (in increasing order) coming from independent realizations of an unknown distribution:

$$x_1 = \left(-\log\frac{38}{40}\right)^2, x_2 = \left(-\log\frac{37}{40}\right)^2, x_3 = \left(-\log\frac{35}{40}\right)^2, x_4 = \left(-\log\frac{34}{40}\right)^2, x_5 = \left(-\log\frac{10}{40}\right)^2.$$

Perform a Kolmogorov-Smirnov test at significance level of 5% to test the null hypothesis that the data given above comes from a Weibull distribution with shape parameter  $\tau = \frac{1}{2}$  and scale parameter c = 1. Moreover, explain why the Kolmogorov-Smirnov test is applicable in this example.